

JAGS Assignment

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4/5/2020

Exercise 8.1

```
data(mtcars)
y = select(mtcars,vs)
s = c( rep("Apple",6) , rep("Blurberry",12) , rep("Cherry",14))
data = data.frame(y=y,s=s)
colnames(data)[1] = c("y")
rownames(data) = c(1:32)
source("DBDA2E-utilities.R")

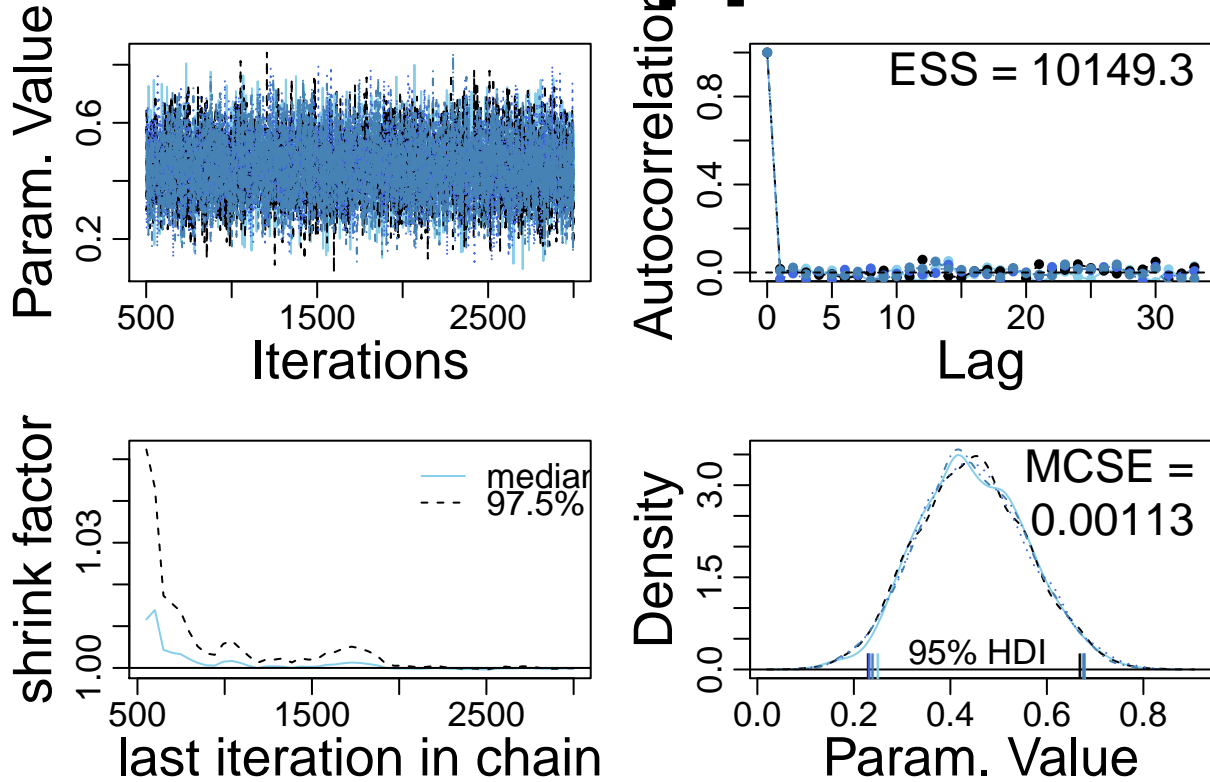
##
## *****
## Kruschke, J. K. (2015). Doing Bayesian Data Analysis, Second Edition:
## A Tutorial with R, JAGS, and Stan. Academic Press / Elsevier.
## *****
source("Jags-Ydich-XnomSsubj-MbernBeta.R")

##
## *****
## Kruschke, J. K. (2015). Doing Bayesian Data Analysis, Second Edition:
## A Tutorial with R, JAGS, and Stan. Academic Press / Elsevier.
## *****
mcmc = genMCMC( data=data , numSavedSteps=10000)

## Compiling model graph
##   Resolving undeclared variables
##   Allocating nodes
## Graph information:
##   Observed stochastic nodes: 32
##   Unobserved stochastic nodes: 3
##   Total graph size: 70
##
## Initializing model
##
## Burning in the MCMC chain...
## Sampling final MCMC chain...

parameterNames = varnames(mcmc)
for ( parName in parameterNames ) {
diagMCMC( codaObject=mcmc , parName=parName )
}
smryMCMC( mcmc , compVal=NULL , compValDiff=0.0 )
```

theta[3]



##	Mean	Median	Mode	ESS	HDImass	
## theta[1]	0.501050655	0.500110989	0.46790615	10000.0	0.95	
## theta[2]	0.438892850	0.436264470	0.43227053	11164.7	0.95	
## theta[3]	0.445870084	0.442474922	0.41903284	10000.0	0.95	
## theta[1]-theta[2]	0.062157804	0.061994723	0.09175565	10000.0	0.95	
## theta[1]-theta[3]	0.055180571	0.059516286	0.08928313	10000.0	0.95	
## theta[2]-theta[3]	-0.006977234	-0.008168533	0.01635828	10000.0	0.95	
##	HDIlow	HDIhigh	CompVal	PcntGtCompVal	ROPElow	ROPEhigh
## theta[1]	0.2150179	0.7894756	NA	NA	NA	NA
## theta[2]	0.2150478	0.6748909	NA	NA	NA	NA
## theta[3]	0.2368439	0.6749915	NA	NA	NA	NA
## theta[1]-theta[2]	-0.3291888	0.4201468	0	62.39	NA	NA
## theta[1]-theta[3]	-0.3031115	0.4219962	0	61.62	NA	NA
## theta[2]-theta[3]	-0.3164877	0.3276956	0	48.10	NA	NA
##	PcntLtROPE	PcntInROPE	PcntGtROPE			
## theta[1]	NA	NA	NA			
## theta[2]	NA	NA	NA			
## theta[3]	NA	NA	NA			
## theta[1]-theta[2]	NA	NA	NA			
## theta[1]-theta[3]	NA	NA	NA			
## theta[2]-theta[3]	NA	NA	NA			
##	Mean	Median	Mode	ESS	HDImass	
## theta[1]	0.501050655	0.500110989	0.46790615	10000.0	0.95	
## theta[2]	0.438892850	0.436264470	0.43227053	11164.7	0.95	
## theta[3]	0.445870084	0.442474922	0.41903284	10000.0	0.95	
## theta[1]-theta[2]	0.062157804	0.061994723	0.09175565	10000.0	0.95	

```
## theta[1]-theta[3] 0.055180571 0.059516286 0.08928313 10000.0 0.95
## theta[2]-theta[3] -0.006977234 -0.008168533 0.01635828 10000.0 0.95
##
## HDIlow HDIhigh CompVal PcntGtCompVal ROPElow ROPEhigh
## theta[1] 0.2150179 0.7894756 NA NA NA NA
## theta[2] 0.2150478 0.6748909 NA NA NA NA
## theta[3] 0.2368439 0.6749915 NA NA NA NA
## theta[1]-theta[2] -0.3291888 0.4201468 0 62.39 NA NA
## theta[1]-theta[3] -0.3031115 0.4219962 0 61.62 NA NA
## theta[2]-theta[3] -0.3164877 0.3276956 0 48.10 NA NA
##
## PcntLtROPE PcntInROPE PcntGtROPE
## theta[1] NA NA NA
## theta[2] NA NA NA
## theta[3] NA NA NA
## theta[1]-theta[2] NA NA NA
## theta[1]-theta[3] NA NA NA
## theta[2]-theta[3] NA NA NA
```

```
plotMCMC( mcmc, data=data , compVal=NULL , compValDiff=0.0 )
```

Looking at the diagonal panels, we see that the HDI for theta[1] is wider than the HDI for theta[2], and the MCSE for theta[2] is wider than the MCSE for theta[3], because the data for theta[1] are far fewer than the data for theta[2].

Exercise 8.2

```
smryMCMC( mcmc , compVal=0.5 , rope=c(0.45,0.55) ,compValDiff=0.0 , ropeDiff = c(-0.05,0.05) )
```

```
##
## Mean Median Mode ESS HDImass
## theta[1] 0.501050655 0.500110989 0.46790615 10000.0 0.95
## theta[2] 0.438892850 0.436264470 0.43227053 11164.7 0.95
## theta[3] 0.445870084 0.442474922 0.41903284 10000.0 0.95
## theta[1]-theta[2] 0.062157804 0.061994723 0.09175565 10000.0 0.95
## theta[1]-theta[3] 0.055180571 0.059516286 0.08928313 10000.0 0.95
## theta[2]-theta[3] -0.006977234 -0.008168533 0.01635828 10000.0 0.95
##
## HDIlow HDIhigh CompVal PcntGtCompVal ROPElow ROPEhigh
## theta[1] 0.2150179 0.7894756 0.5 50.04 0.45 0.55
## theta[2] 0.2150478 0.6748909 0.5 30.60 0.45 0.55
## theta[3] 0.2368439 0.6749915 0.5 31.52 0.45 0.55
## theta[1]-theta[2] -0.3291888 0.4201468 0.0 62.39 -0.05 0.05
## theta[1]-theta[3] -0.3031115 0.4219962 0.0 61.62 -0.05 0.05
## theta[2]-theta[3] -0.3164877 0.3276956 0.0 48.10 -0.05 0.05
##
## PcntLtROPE PcntInROPE PcntGtROPE
## theta[1] 37.66 24.17 38.17
## theta[2] 54.52 27.16 18.32
## theta[3] 52.78 28.87 18.35
## theta[1]-theta[2] 28.44 19.02 52.54
## theta[1]-theta[3] 29.68 18.72 51.60
## theta[2]-theta[3] 40.52 23.32 36.16
##
## Mean Median Mode ESS HDImass
## theta[1] 0.501050655 0.500110989 0.46790615 10000.0 0.95
## theta[2] 0.438892850 0.436264470 0.43227053 11164.7 0.95
## theta[3] 0.445870084 0.442474922 0.41903284 10000.0 0.95
## theta[1]-theta[2] 0.062157804 0.061994723 0.09175565 10000.0 0.95
## theta[1]-theta[3] 0.055180571 0.059516286 0.08928313 10000.0 0.95
```

```
## theta[2]-theta[3] -0.006977234 -0.008168533 0.01635828 10000.0 0.95
##          HDIlow  HDIhigh CompVal PcntGtCompVal ROPElow ROPEhigh
## theta[1]      0.2150179 0.7894756 0.5          50.04 0.45 0.55
## theta[2]      0.2150478 0.6748909 0.5          30.60 0.45 0.55
## theta[3]      0.2368439 0.6749915 0.5          31.52 0.45 0.55
## theta[1]-theta[2] -0.3291888 0.4201468 0.0          62.39 -0.05 0.05
## theta[1]-theta[3] -0.3031115 0.4219962 0.0          61.62 -0.05 0.05
## theta[2]-theta[3] -0.3164877 0.3276956 0.0          48.10 -0.05 0.05
##          PcntLtROPE PcntInROPE PcntGtROPE
## theta[1]          37.66      24.17      38.17
## theta[2]          54.52      27.16      18.32
## theta[3]          52.78      28.87      18.35
## theta[1]-theta[2]      28.44      19.02      52.54
## theta[1]-theta[3]      29.68      18.72      51.60
## theta[2]-theta[3]      40.52      23.32      36.16
```

Each row represents the parameter or parameter difference. The columns labelled Mean, Median, and Mode are the corresponding values of parameter. The ESS is the effective sample size. The next three columns indicate the probability mass of the HDI, the lower limit of the HDI, and the upper limit. The comparison value is specified in the argument as 0.5. The next column indicates the percentage of the posterior that is greater than the comparison value (PcntGtCompVal). The ROPE columns repeat the specifications in the arguments. The final three columns indicate the percentage of the posterior distribution that is less than the ROPE lower limit, within the ROPE limits, and greater than the ROPE upper limit.

Exercise 8.3

```
##The first line above specifies the beginning of the filenames for saved information
##The second line above specifies the graphics format for saved graphs

fileNameRoot = "Jags-Ydich-XnomSsubj-MbernBeta-Yiru-"
graphFileType = "eps"

##The MCMC chain is saved in a file named Jags-Ydich-XnomSsubj-MbernBeta-Yiru-Mcmc.Rdata
mcmc = genMCMC( data=data , numSavedSteps=10000 , saveName=fileNameRoot )

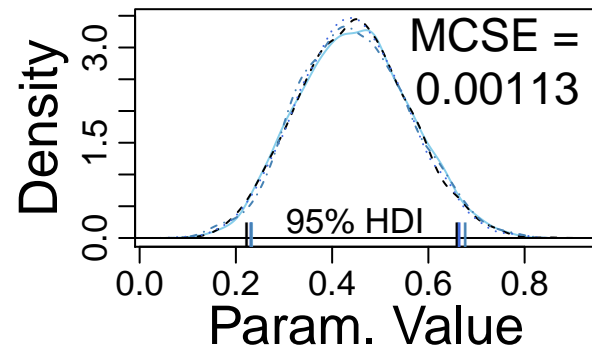
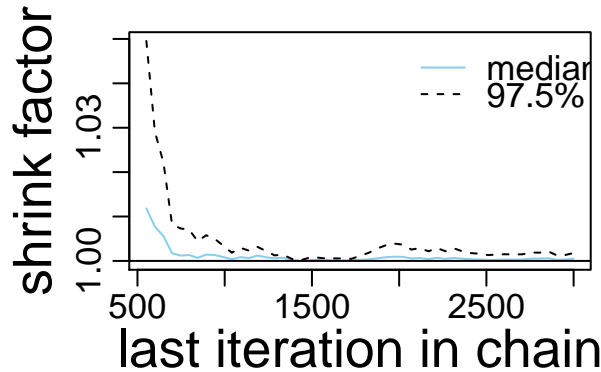
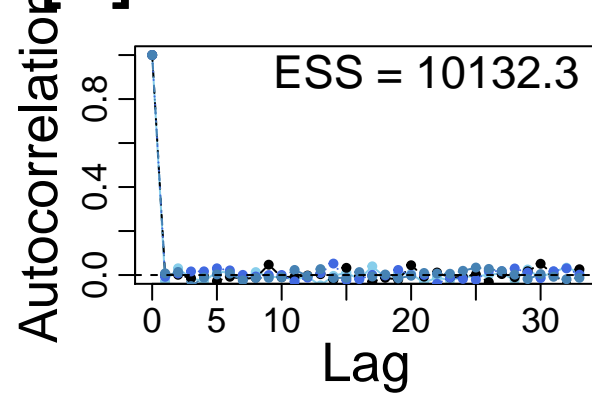
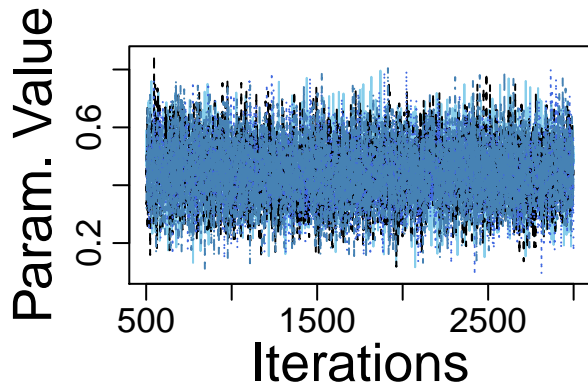
## Compiling model graph
##   Resolving undeclared variables
##   Allocating nodes
## Graph information:
##   Observed stochastic nodes: 32
##   Unobserved stochastic nodes: 3
##   Total graph size: 70
##
## Initializing model
##
## Burning in the MCMC chain...
## Sampling final MCMC chain...

##The diagnostic graphs are saved in files named Jags-Ydich-XnomSsubj-MbernBeta-Yiru - Diagtheta.eps
parameterNames = varnames(mcmc)
for ( parName in parameterNames ) {
diagMCMC( codaObject=mcmc , parName=parName , saveName=fileNameRoot , saveType=graphFileType )
}

##The summary information is saved in a file named Jags-Ydich-XnomSsubj-MbernBeta-Yiru-detail.csv
```

```
detail = smryMCMC(mcmc, compVal=0.5, rope=c(0.45,0.55),compValDiff=0.0, ropeDiff = c(-0.05,0.05) ,saveName=
```

theta[3]



	Mean	Median	Mode	ESS	HDImass
## theta[1]	0.498830371	0.498094745	0.46635439	9513.6	0.95
## theta[2]	0.437224701	0.435861151	0.44300529	10000.0	0.95
## theta[3]	0.444421319	0.442593500	0.44206287	10000.0	0.95
## theta[1]-theta[2]	0.061605670	0.063743891	0.09710688	10000.0	0.95
## theta[1]-theta[3]	0.054409051	0.056214744	0.09552906	10000.0	0.95
## theta[2]-theta[3]	-0.007196618	-0.008737934	-0.01532545	10000.0	0.95

	HDILow	HDIhigh	CompVal	PcntGtCompVal	ROPElow	ROPEhigh
## theta[1]	0.1990578	0.7765479	0.5	49.59	0.45	0.55
## theta[2]	0.2060644	0.6664845	0.5	29.98	0.45	0.55
## theta[3]	0.2311865	0.6694458	0.5	31.05	0.45	0.55
## theta[1]-theta[2]	-0.3149150	0.4300225	0.0	62.59	-0.05	0.05
## theta[1]-theta[3]	-0.3193779	0.4173329	0.0	60.67	-0.05	0.05
## theta[2]-theta[3]	-0.3264288	0.3133183	0.0	47.92	-0.05	0.05

	PcntLtROPE	PcntInROPE	PcntGtROPE
## theta[1]	38.59	23.28	38.13
## theta[2]	54.84	26.76	18.40
## theta[3]	52.54	29.36	18.10
## theta[1]-theta[2]	28.38	18.96	52.66
## theta[1]-theta[3]	30.17	18.94	50.89
## theta[2]-theta[3]	40.25	23.09	36.66

```
##The graph of the posterior distribution is saved in a file named Jags-Ydich-XnomSsubj-MbernBeta-Yiru-
plotMCMC( mcmc, data=data , compVal=NULL, compValDiff=0.0, saveName=fileNameRoot , saveType=graphFileTy
```

```
## pdf
## 2
```

Exercise 8.4

(A)

```
source("Jags-Ydich-XnomSsubj-MbernBeta-8.4.R")

##
## *****
## Kruschke, J. K. (2015). Doing Bayesian Data Analysis, Second Edition:
## A Tutorial with R, JAGS, and Stan. Academic Press / Elsevier.
## *****

fileNameRoot = "Jags-Ydich-XnomSsubj-MbernBeta-Yiru8.4-"
graphFileType = "eps"

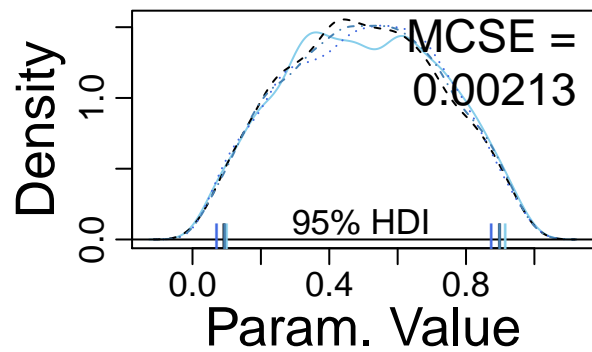
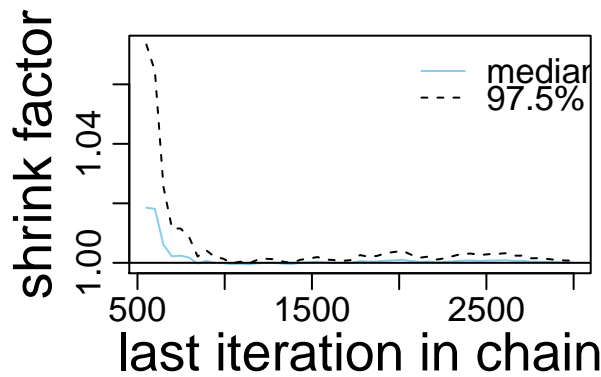
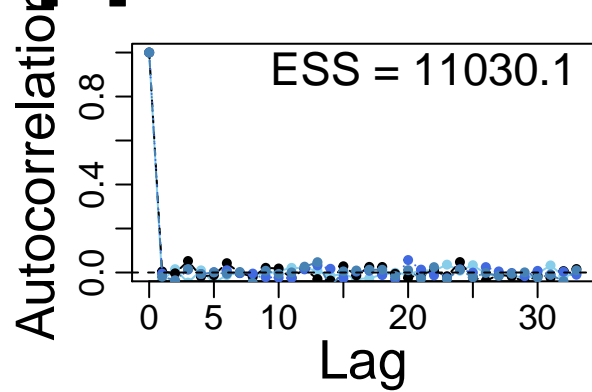
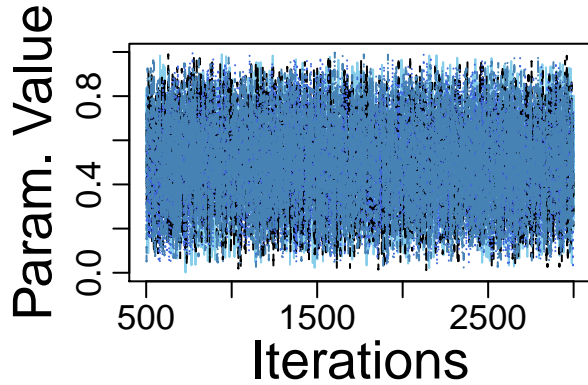
##The MCMC chain is saved in a file named Jags-Ydich-XnomSsubj-MbernBeta-Yiru8.4-Mcmc.Rdata
mcmc = genMCMC_8.4( data=data , numSavedSteps=10000 , saveName=fileNameRoot )

## Compiling model graph
##   Resolving undeclared variables
##   Allocating nodes
## Graph information:
##   Observed stochastic nodes: 0
##   Unobserved stochastic nodes: 35
##   Total graph size: 70
##
## Initializing model
##
## Burning in the MCMC chain...
## Sampling final MCMC chain...

##The diagnostic graphs are saved in files named Jags-Ydich-XnomSsubj-MbernBeta-Yiru8.4 - Diagtheta.eps
parameterNames = varnames(mcmc)
for ( parName in parameterNames ) {
diagMCMC( codaObject=mcmc , parName=parName , saveName=fileNameRoot , saveType=graphFileType )
}

##The summary information is saved in a file named Jags-Ydich-XnomSsubj-MbernBeta-Yiru8.4-detail.csv
detail = smryMCMC(mcmc, compVal=0.5, rope=c(0.45,0.55),compValDiff=0.0, ropeDiff = c(-0.05,0.05) ,saveName=
```

theta[3]



##	Mean	Median	Mode	ESS	HDI	mass
## theta[1]	0.492933818	0.492339150	0.44432969	10000.0	0.95	
## theta[2]	0.497094810	0.498600540	0.54241812	10000.0	0.95	
## theta[3]	0.500284222	0.501128426	0.46326029	10605.9	0.95	
## theta[1]-theta[2]	-0.004160992	-0.005562739	0.02443080	10000.0	0.95	
## theta[1]-theta[3]	-0.007350404	-0.009568616	-0.04312493	10000.0	0.95	
## theta[2]-theta[3]	-0.003189412	-0.002024378	0.03771356	10000.0	0.95	
##	HDIlow	HDIhigh	CompVal	PcntGtCompVal	ROPElow	ROPEhigh
## theta[1]	0.09359287	0.8912644	0.5	48.81	0.45	0.55
## theta[2]	0.07942574	0.8978037	0.5	49.81	0.45	0.55
## theta[3]	0.08814318	0.8990942	0.5	50.11	0.45	0.55
## theta[1]-theta[2]	-0.60987098	0.6002949	0.0	49.33	-0.05	0.05
## theta[1]-theta[3]	-0.61780055	0.5823259	0.0	48.99	-0.05	0.05
## theta[2]-theta[3]	-0.63735826	0.5835598	0.0	49.72	-0.05	0.05
##	PcntLtROPE	PcntInROPE	PcntGtROPE			
## theta[1]	43.66	15.03	41.31			
## theta[2]	42.83	14.92	42.25			
## theta[3]	42.29	14.69	43.02			
## theta[1]-theta[2]	44.81	11.94	43.25			
## theta[1]-theta[3]	44.98	11.83	43.19			
## theta[2]-theta[3]	44.33	11.85	43.82			

##The graph of the posterior distribution is saved in a file named Jags-Ydich-XnomSsubj-MbernBeta-Yiru8
 plotMCMC(mcmc, data=data , compVal=NULL, compValDiff=0.0, saveName=fileNameRoot , saveType=graphFileType)

pdf
 ## 2

(B)

```
source("Jags-Ydich-XnomSsubj-MbernBeta-b.R")

##
## *****
## Kruschke, J. K. (2015). Doing Bayesian Data Analysis, Second Edition:
## A Tutorial with R, JAGS, and Stan. Academic Press / Elsevier.
## *****

fileNameRoot = "Jags-Ydich-XnomSsubj-MbernBeta-YiruB-"
graphFileType = "eps"

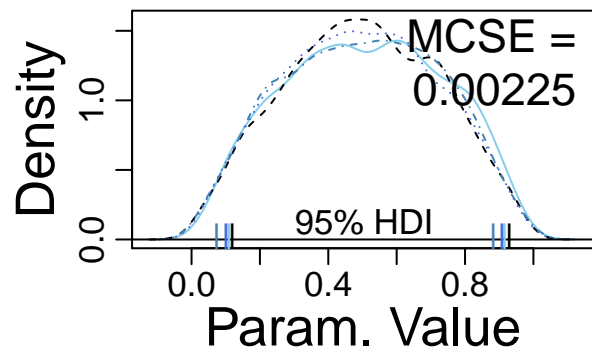
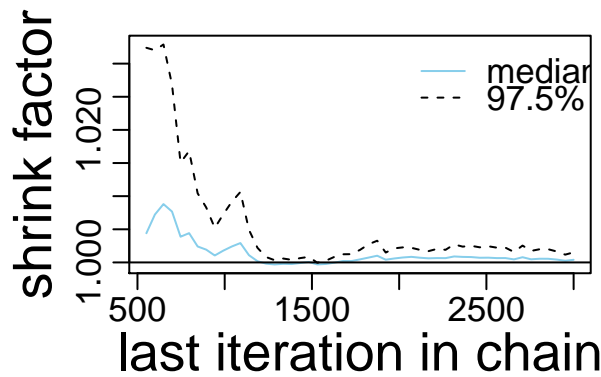
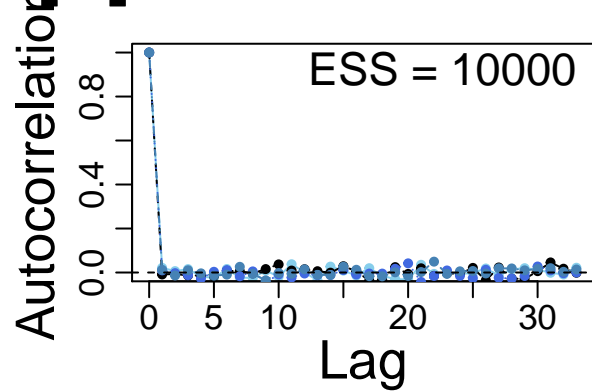
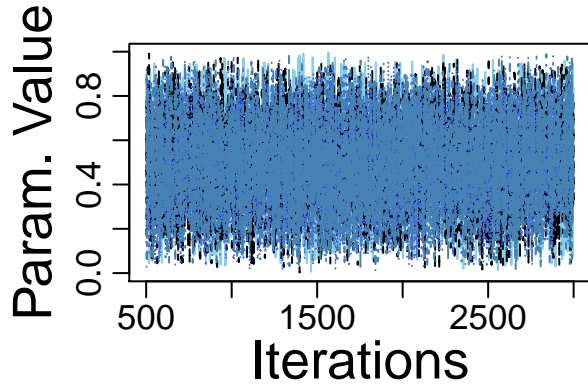
##The MCMC chain is saved in a file named Jags-Ydich-XnomSsubj-MbernBeta-YiruB-Mcmc.Rdata
mcmc = genMCMC_8.4( data=data , numSavedSteps=10000 , saveName=fileNameRoot )

## Compiling model graph
##   Resolving undeclared variables
##   Allocating nodes
## Graph information:
##   Observed stochastic nodes: 0
##   Unobserved stochastic nodes: 35
##   Total graph size: 70
##
## Initializing model
##
## Burning in the MCMC chain...
## Sampling final MCMC chain...

##The diagnostic graphs are saved in files named Jags-Ydich-XnomSsubj-MbernBeta-YiruB - Diagtheta.eps
parameterNames = varnames(mcmc)
for ( parName in parameterNames ) {
diagMCMC( codaObject=mcmc , parName=parName , saveName=fileNameRoot , saveType=graphFileType )
}

##The summary information is saved in a file named Jags-Ydich-XnomSsubj-MbernBeta-YiruB-detail.csv
detail = smryMCMC(mcmc, compVal=0.5, rope=c(0.45,0.55),compValDiff=0.0, ropeDiff = c(-0.05,0.05) ,saveName=
```


theta[3]



##		Mean	Median	Mode	ESS	HDI	mass
##	theta[1]	0.5016132135	0.503419082	0.51047756	10000.0		0.95
##	theta[2]	0.5018419905	0.503223807	0.50659554	10000.0		0.95
##	theta[3]	0.4993229361	0.498385513	0.45641716	10000.0		0.95
##	theta[1]-theta[2]	-0.0002287771	-0.001612801	-0.02884904	10366.9		0.95
##	theta[1]-theta[3]	0.0022902774	0.004093001	0.04554832	10000.0		0.95
##	theta[2]-theta[3]	0.0025190545	0.003014846	0.07624786	10000.0		0.95
##		HDIlow	HDIhigh	CompVal	PcntGtCompVal	ROPElow	ROPEhigh
##	theta[1]	0.09045638	0.9018402	0.5	50.61	0.45	0.55
##	theta[2]	0.09937614	0.9092413	0.5	50.53	0.45	0.55
##	theta[3]	0.09645222	0.9104426	0.5	49.81	0.45	0.55
##	theta[1]-theta[2]	-0.61092649	0.5908724	0.0	49.79	-0.05	0.05
##	theta[1]-theta[3]	-0.58376112	0.6342119	0.0	50.49	-0.05	0.05
##	theta[2]-theta[3]	-0.60396402	0.6170514	0.0	50.37	-0.05	0.05
##		PcntLtROPE	PcntInROPE	PcntGtROPE			
##	theta[1]	41.86	15.46	42.68			
##	theta[2]	42.02	15.02	42.96			
##	theta[3]	42.55	14.56	42.89			
##	theta[1]-theta[2]	44.23	12.09	43.68			
##	theta[1]-theta[3]	43.85	11.81	44.34			
##	theta[2]-theta[3]	43.91	11.11	44.98			

##The graph of the posterior distribution is saved in a file named Jags-Ydich-XnomSsubj-MbernBeta-YiruB
 plotMCMC(mcmc, data=data , compVal=NULL, compValDiff=0.0, saveName=fileNameRoot , saveType=graphFileType)

pdf
 ## 2

The distributions on theta[1], theta[2] and theta[3] look uniform.

(C)

```
source("Jags-Ydich-XnomSsubj-MbernBeta-c.R")

##
## *****
## Kruschke, J. K. (2015). Doing Bayesian Data Analysis, Second Edition:
## A Tutorial with R, JAGS, and Stan. Academic Press / Elsevier.
## *****

fileNameRoot = "Jags-Ydich-XnomSsubj-MbernBeta-YiruC-"
graphFileType = "eps"

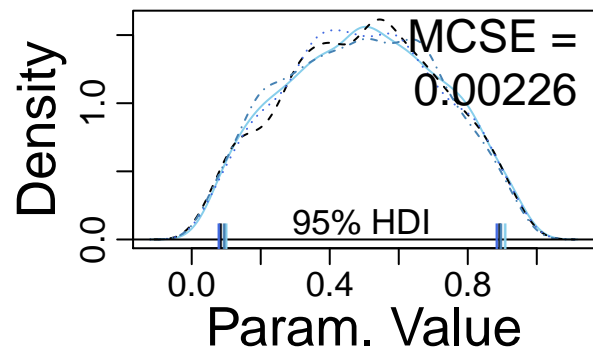
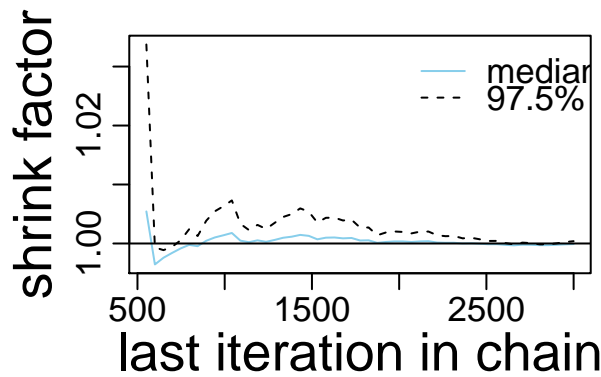
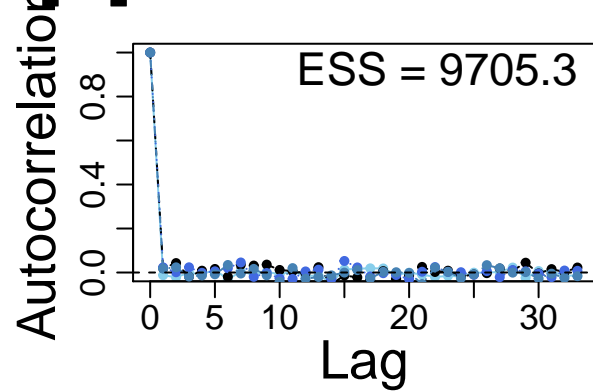
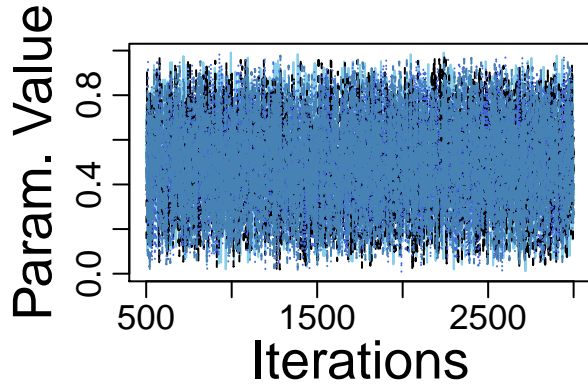
##The MCMC chain is saved in a file named Jags-Ydich-XnomSsubj-MbernBeta-YiruC-Mcmc.Rdata
mcmc = genMCMC_8.4( data=data , numSavedSteps=10000 , saveName=fileNameRoot )

## Compiling model graph
##   Resolving undeclared variables
##   Allocating nodes
## Graph information:
##   Observed stochastic nodes: 0
##   Unobserved stochastic nodes: 35
##   Total graph size: 70
##
## Initializing model
##
## Burning in the MCMC chain...
## Sampling final MCMC chain...

##The diagnostic graphs are saved in files named Jags-Ydich-XnomSsubj-MbernBeta-YiruC - Diagtheta.eps
parameterNames = varnames(mcmc)
for ( parName in parameterNames ) {
diagMCMC( codaObject=mcmc , parName=parName , saveName=fileNameRoot , saveType=graphFileType )
}

##The summary information is saved in a file named Jags-Ydich-XnomSsubj-MbernBeta-YiruC-detail.csv
detail = smryMCMC(mcmc, compVal=0.5, rope=c(0.45,0.55),compValDiff=0.0, ropeDiff = c(-0.05,0.05) ,saveName=
```

theta[3]



##	Mean	Median	Mode	ESS	HDI	mass
## theta[1]	0.5024123575	0.502071002	0.46958592	10352.8	0.95	
## theta[2]	0.4963250604	0.498354798	0.55983639	10000.0	0.95	
## theta[3]	0.4959867428	0.499375071	0.52246229	10000.0	0.95	
## theta[1]-theta[2]	0.0060872971	0.007739527	0.02070282	10000.0	0.95	
## theta[1]-theta[3]	0.0064256147	0.012424091	0.06828943	9599.1	0.95	
## theta[2]-theta[3]	0.0003383176	-0.005061303	-0.05756460	10000.0	0.95	
##	HDIlow	HDIhigh	CompVal	PcntGtCompVal	ROPElow	ROPEhigh
## theta[1]	0.10239081	0.9054275	0.5	50.33	0.45	0.55
## theta[2]	0.09043379	0.9041781	0.5	49.76	0.45	0.55
## theta[3]	0.07856499	0.8865914	0.5	49.91	0.45	0.55
## theta[1]-theta[2]	-0.59011956	0.6034972	0.0	50.90	-0.05	0.05
## theta[1]-theta[3]	-0.62010338	0.5798768	0.0	51.33	-0.05	0.05
## theta[2]-theta[3]	-0.61527133	0.6117624	0.0	49.48	-0.05	0.05
##	PcntLtROPE	PcntInROPE	PcntGtROPE			
## theta[1]	42.22	14.82	42.96			
## theta[2]	43.07	14.83	42.10			
## theta[3]	42.69	15.29	42.02			
## theta[1]-theta[2]	42.95	12.31	44.74			
## theta[1]-theta[3]	43.18	11.61	45.21			
## theta[2]-theta[3]	44.37	12.07	43.56			

##The graph of the posterior distribution is saved in a file named Jags-Ydich-XnomSsubj-MbernBeta-YiruC
 plotMCMC(mcmc, data=data , compVal=NULL, compValDiff=0.0, saveName=fileNameRoot , saveType=graphFileType)

pdf
 ## 2

The distributions on $\theta[1]$, $\theta[2]$ and $\theta[3]$ look uniform.